

Southeastern United States Summer Rainfall Framework and its Implication for Seasonal Forecast

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Motivations

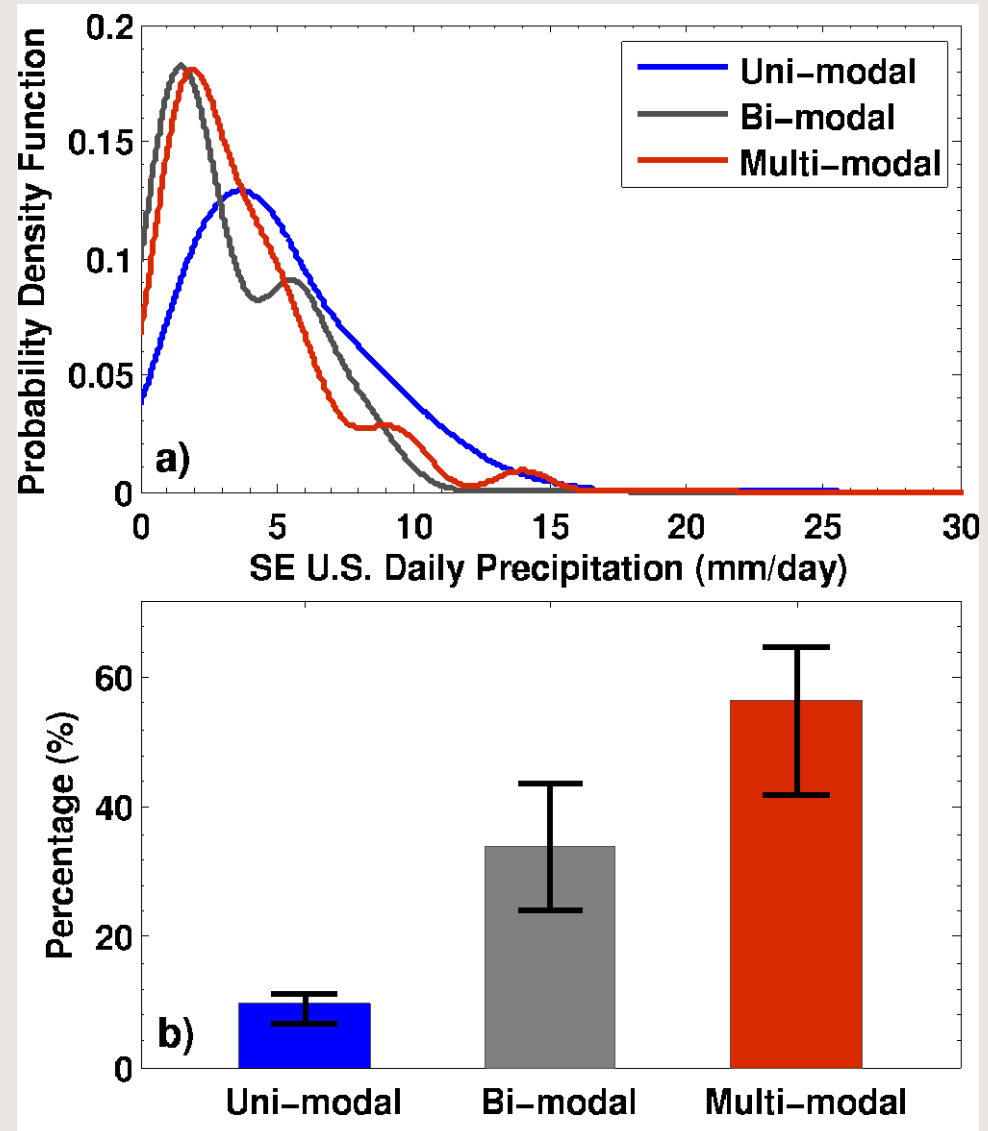
- The Southeastern United States (SE US) is one of the fastest developing regions in the nation, thus the warm season precipitation becomes increasingly more important for this region.
- Reliable climate prediction for SE US relies on accurate statistical inference on summer rainfall.



2005 SE US Flood

Motivations

- Previous rainfall statistical inference takes the approach of fitting probability distributions with predefined kernels (e.g. Gamma, Weibull).
- These traditional distribution fail to capture the “multi-modal” features of SE US summer rainfall; thus cannot provide reliable statistical inference.



Goal of This Study

- Construct a new probability framework that can well represent the “multi-modal” features of SE US summer rainfall distribution
- Explore the possible linkages between climate factors and the complete spectrum of rainfall distribution
- Assess the potential predictability skill according to these linkages

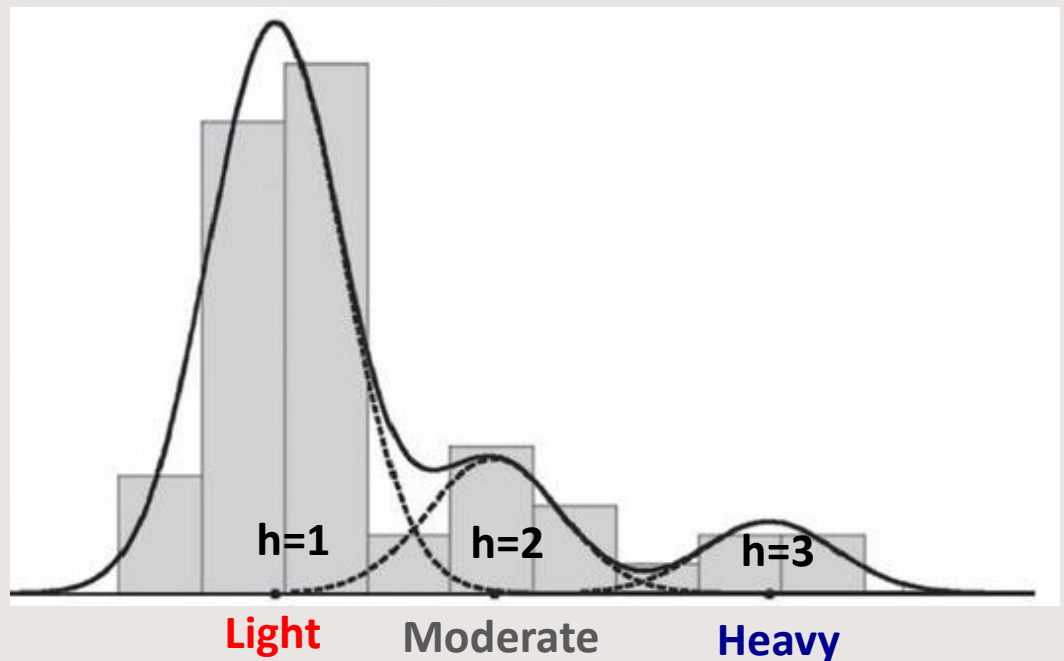
Configuration of Rainfall Framework

3-cluster *Normal* Mixture Models:

$$y_i | \pi, \mu, \phi \sim \sum_{h=1}^3 \pi_h N(y_i | \mu_h, \phi_h)$$

3 rainfall clusters corresponding to **light** (h=1), moderate (h=2), and **heavy** (h=3) rainfall

π_h : Cluster weight
 μ_h : Cluster mean
 Φ_h : Cluster precision
(variance⁻¹)



Bayesian Inference

- Prior Distribution (Expert knowledge but weakly informative)

$$\rho_h \sim \text{Dirichlet}(a_1, a_2, a_3) \quad a_1 = 0.5; a_2 = 0.35; a_3 = 0.15$$

$$(m_h, f_h) \sim \text{Normal}(m_h | m_{0h}, K t_h) \text{Gamma}(t_h | a_{th}, b_{th})$$

$$m_{0h} = (1.0, 8.0, 20.0) \quad \text{AMS definition of rainfall types}$$

$$K = (1, 1, 1); a_{th} = (1.0, 1.0, 0.4); b_{th} = (1.0, 1.0, 1.0) \quad \text{Weakly informative}$$

- Full Conditional Posteriors

$$(m_h, f_h | -) \sim N(m_h | \hat{m}_h, \hat{K} f_h) \text{Ga}(f_h | \hat{a}_{th}, \hat{b}_{th})$$

$$\hat{K} = (K^{-1} + n_h)^{-1}; \hat{m}_h = \hat{K} (K^{-1} m_0 + n_h \bar{y}_h); \hat{a}_{th} = a_{th} + n_h/2$$

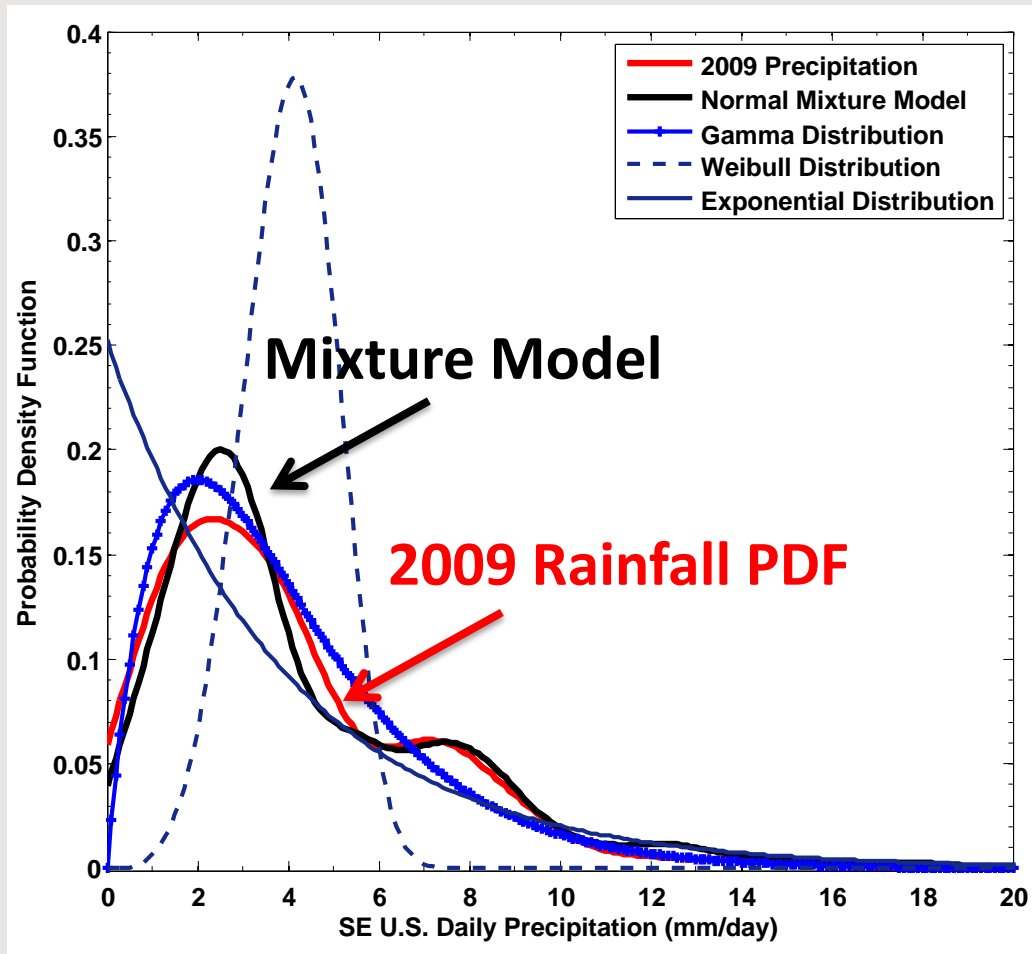
$$\hat{b}_{th} = b_t + \frac{1}{2} \left\{ \sum_{i=1}^{n_h} (y_i - \bar{y}_h)^2 + \left(\frac{n_h}{1 + K n_h} \right) (\bar{y}_h - m_0)^2 \right\}$$

$$(\rho_1, \rho_2, \rho_3 | -) \sim \text{Dirichlet}(a_1 + n_1, a_2 + n_2, a_3 + n_3)$$

Gibbs Sampler

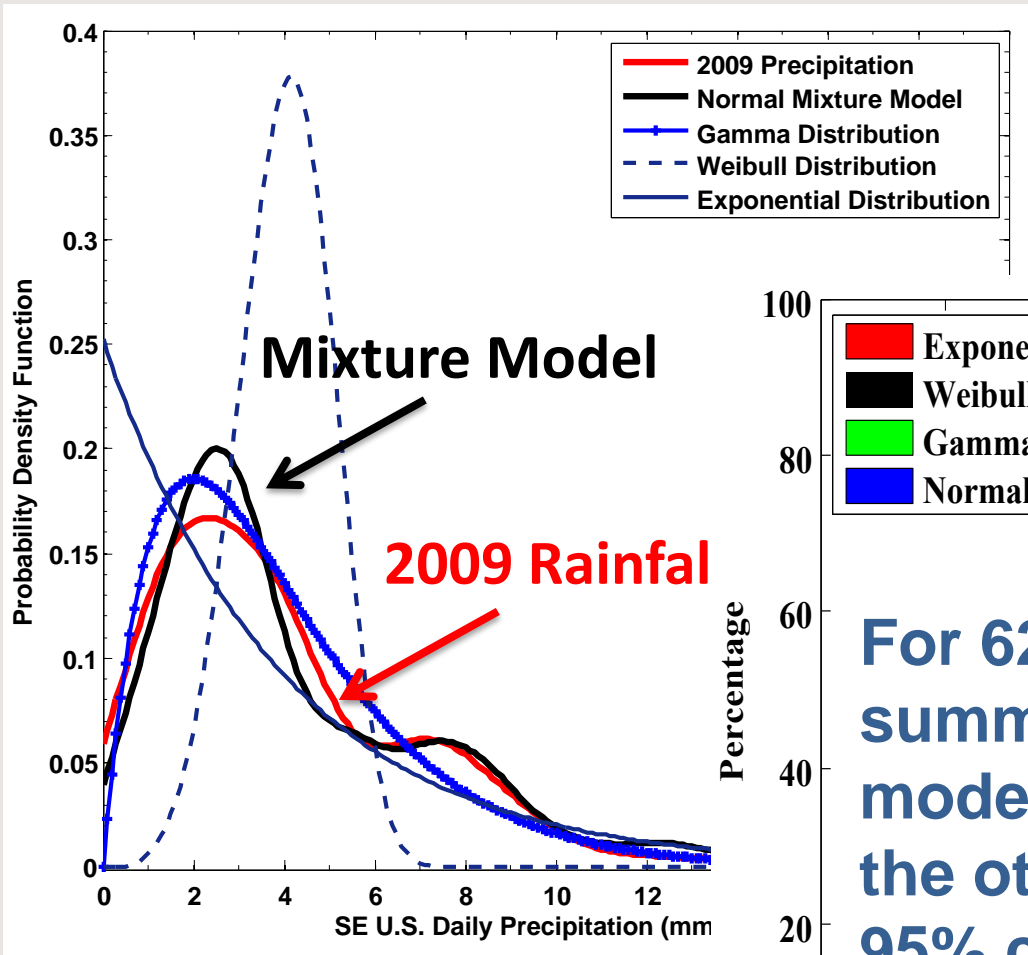
- Markov Chain Monte Carlo (MCMC)
 - Update cluster members from likelihood model
 - Update $(m_h, f_h | -) \sim N(m_h | \hat{m}_h, \hat{\sigma}_h^2) Ga(f_h | \hat{a}_{th}, \hat{b}_{th})$
 - Update $(\rho_1, \rho_2, \rho_3 | -) \sim Dirichlet(a_1 + n_1, a_2 + n_2, a_3 + n_3)$
- Constraints on μ_h ($\mu_1 < \mu_2 < \mu_3$) to deal with label switching problems
- Run 1000-iteration MCMC, the 1st 200 burn-in samples are discarded

Model Performance: 2009 case

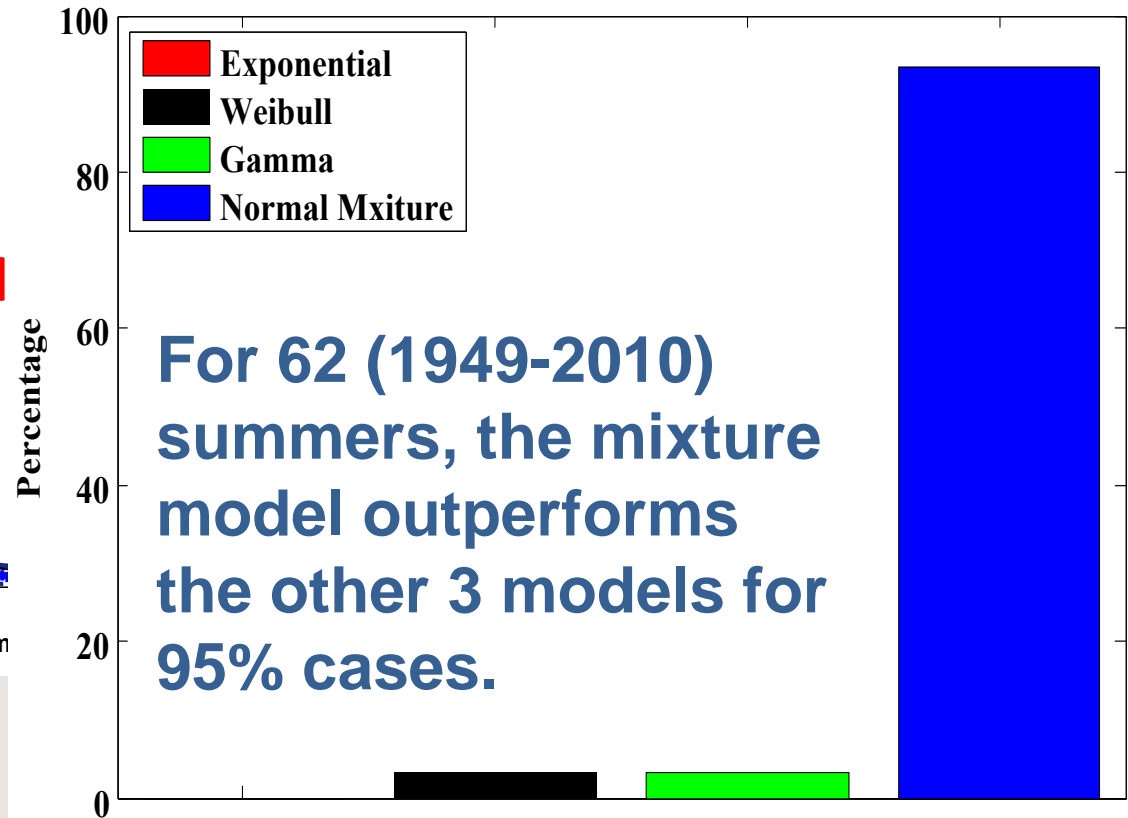


3-cluster *Normal* Mixture model outperforms traditional distribution models in describing the probability behavior of SE US summer rainfall

Model Performance: Goodness of Fit

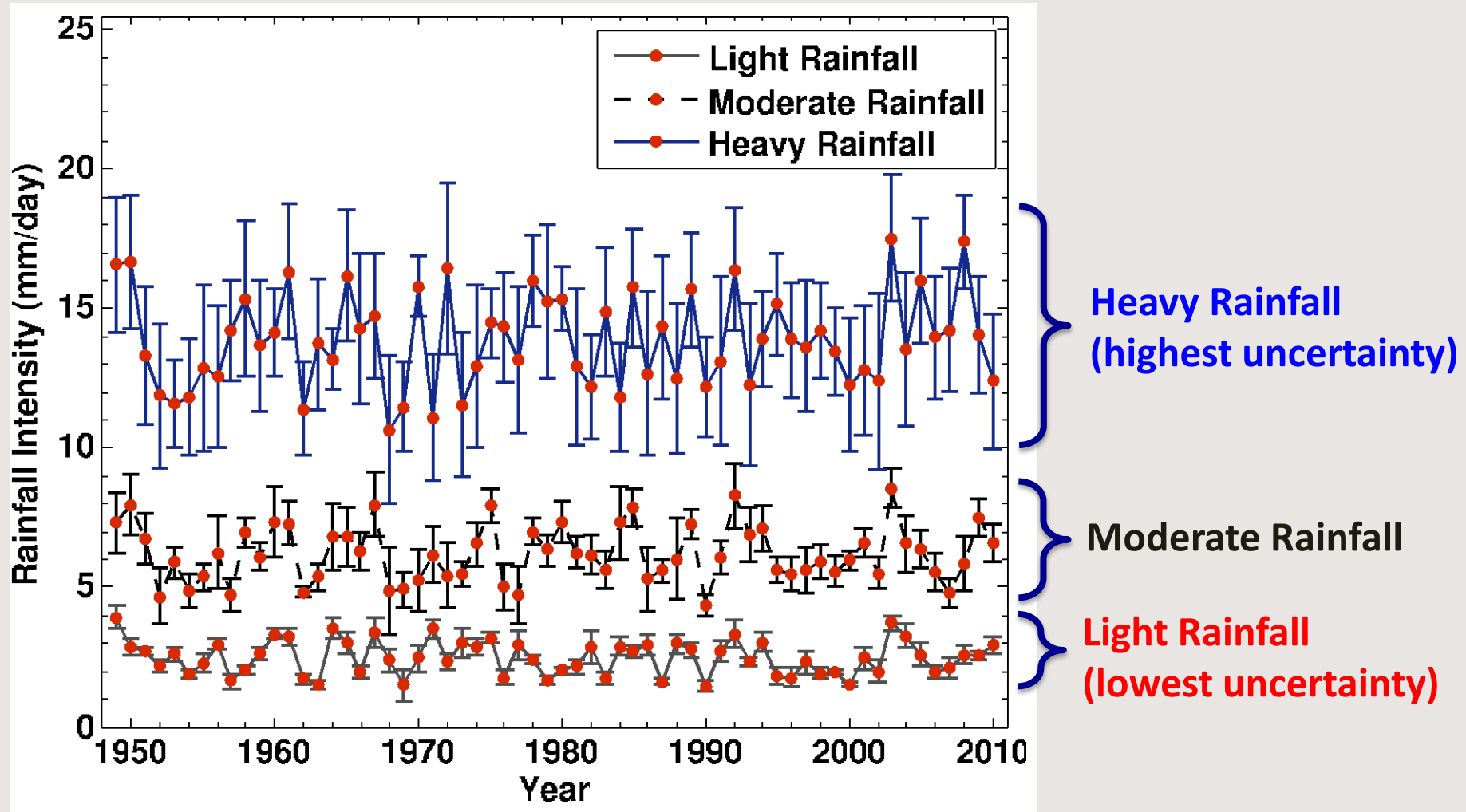


3-cluster *Normal* Mixture model outperforms traditional

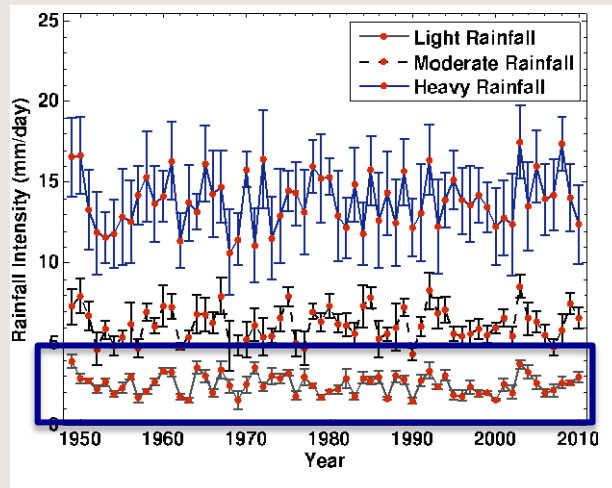


For 62 (1949-2010) summers, the mixture model outperforms the other 3 models for 95% cases.

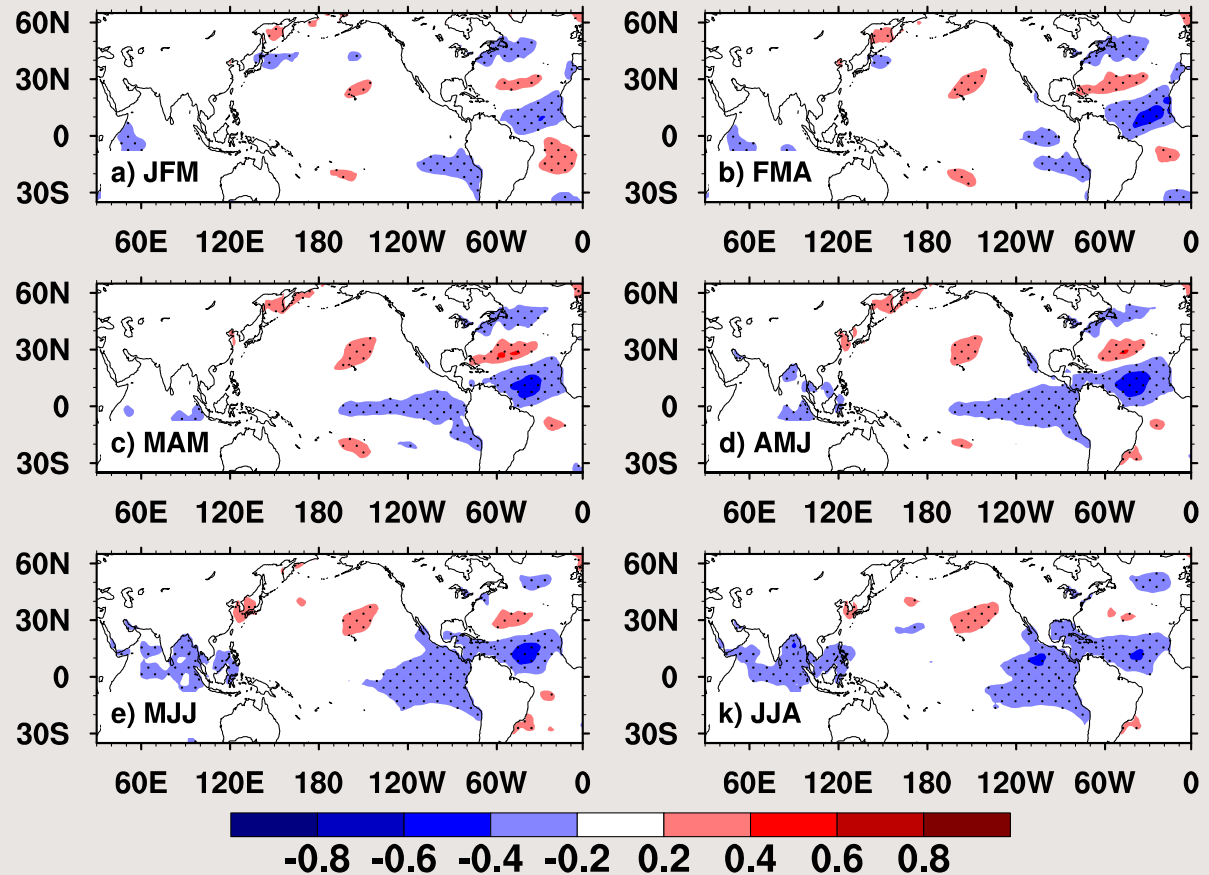
Bayesian Inference on μ_h



Light Rainfall & SSTA

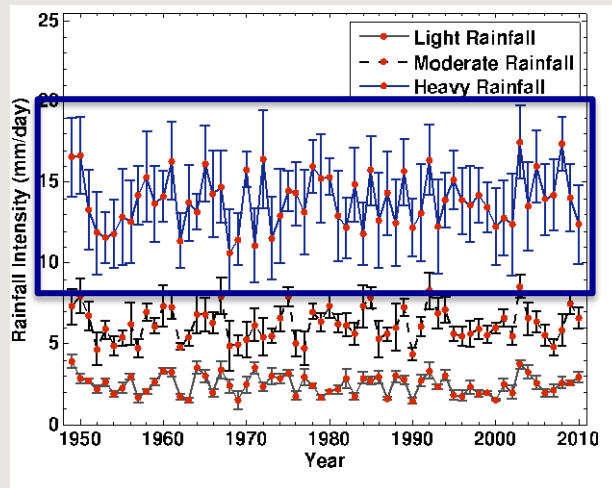


**La Niña and tripole
SSTA over the North
Atlantic occur 3
months ahead of the
SE US light rainfall**

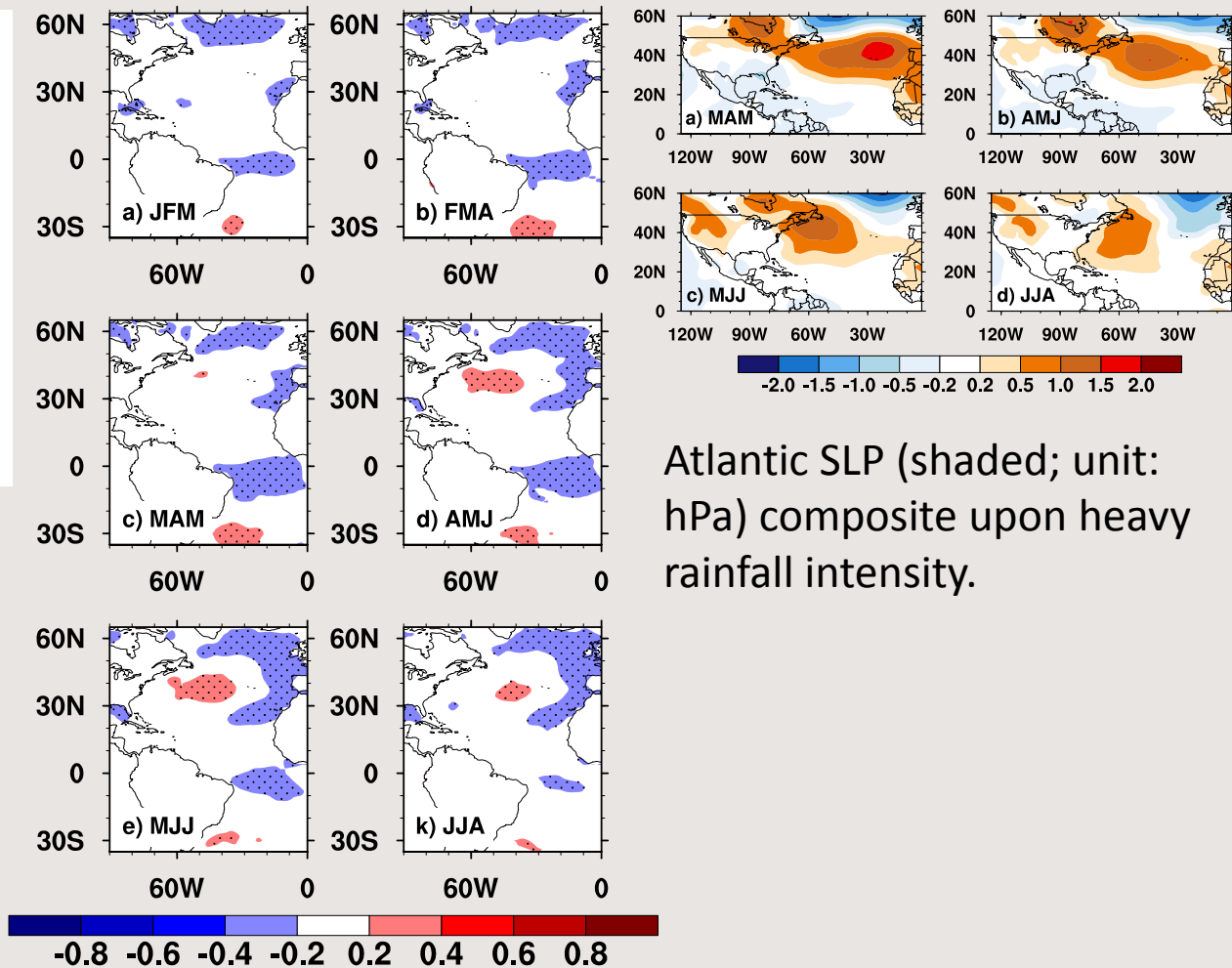


Global SSTA (shaded; unit: normalized; dimensionless) regressed upon light rainfall intensity. SSTA significant at 0.05 level by 1000-time Monte-Carlo simulation is stippled.

Heavy Rainfall & SSTA



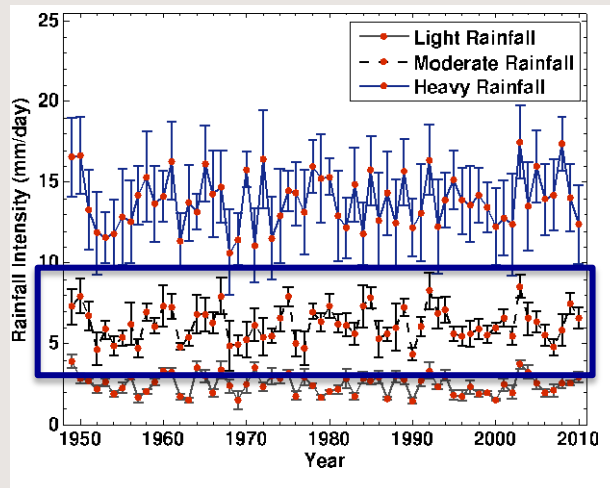
SE US Heavy rainfall is associated with a “horseshoe-like” SSTA over the North Atlantic.



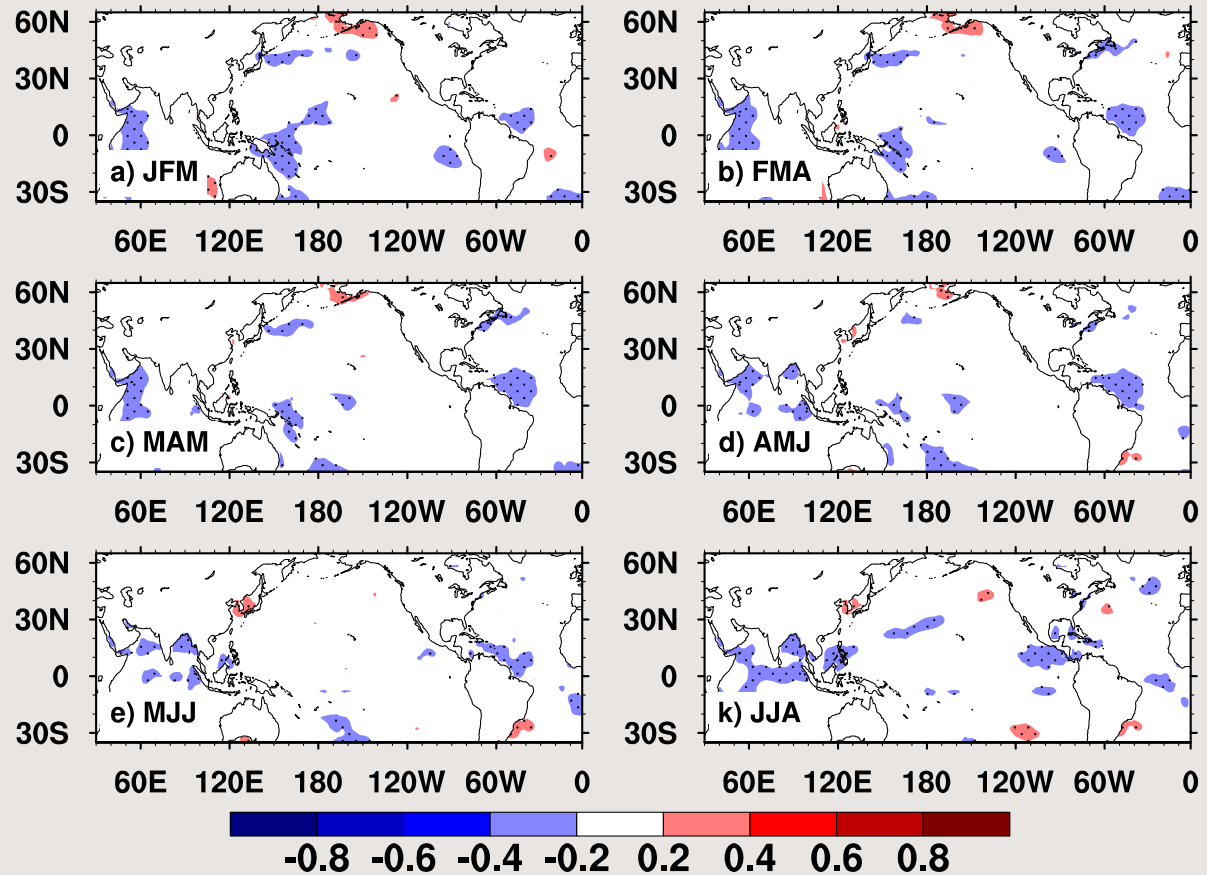
Atlantic SLP (shaded; unit: hPa) composite upon heavy rainfall intensity.

Atlantic SSTA (shaded; unit: normalized; dimensionless) regressed upon heavy rainfall intensity. SSTA significant at 0.05 level by 1000-time Monte-Carlo simulation is stippled.

Moderate Rainfall & SSTA



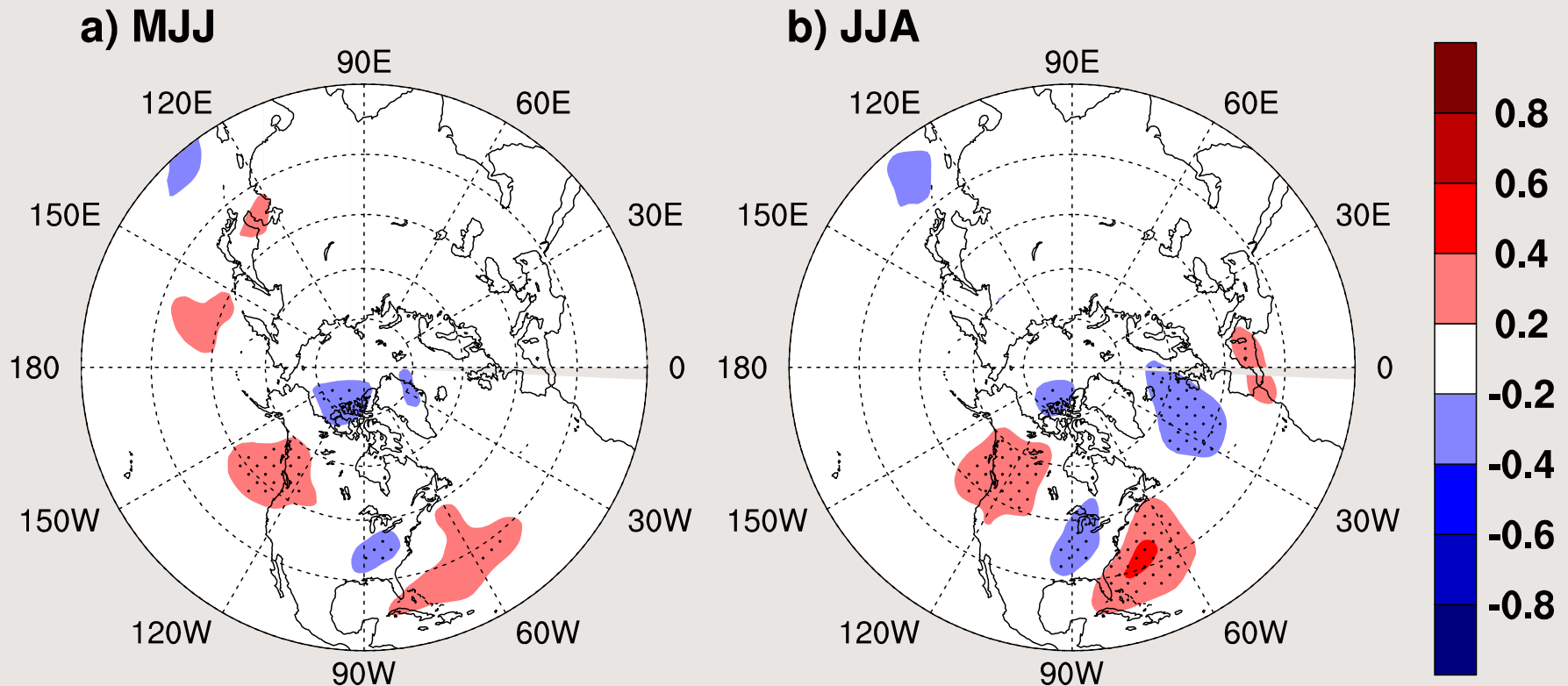
**No systematic
SSTA pattern
concurs with
moderate rainfall**



Global SSTA (shaded; unit: normalized; dimensionless) regressed upon moderate rainfall intensity. SSTA significant at 0.05 level by 1000-time Monte-Carlo simulation is stippled.

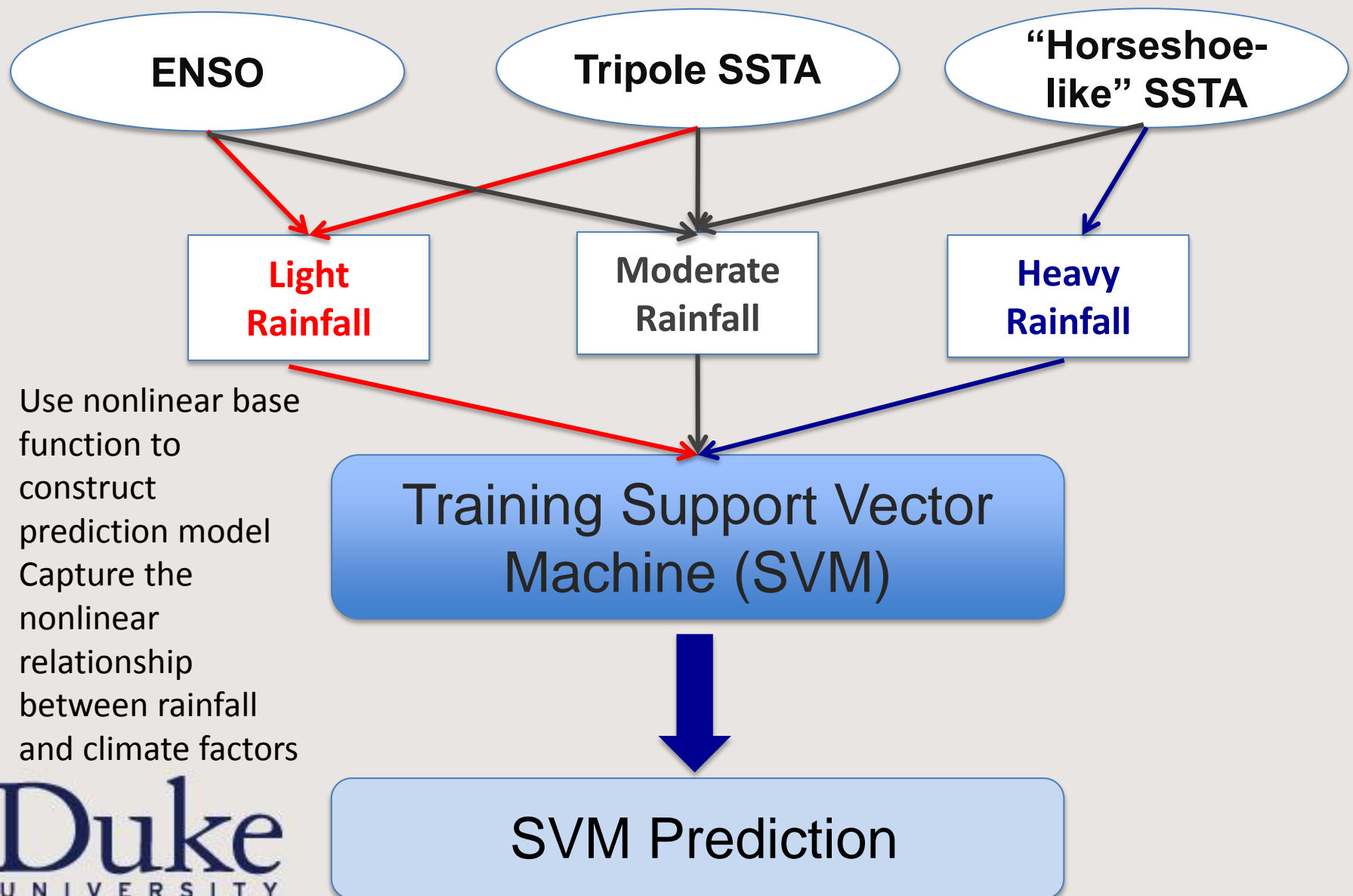
Moderate Rainfall & Atmo. Circulation

Moderate rainfall is likely generated due to atmospheric internal variability



a) MJJ and b) JJA 850hPa geopotential height anomaly (shaded; Units: normalized, dimensionless) regressed upon SE US moderate rainfall intensity. Geopotential height anomalies significant at level by 1000-time Monte-Carlo simulation are stippled (From Li and Li, 2013: *Environ. Res. Letts.* In Press)

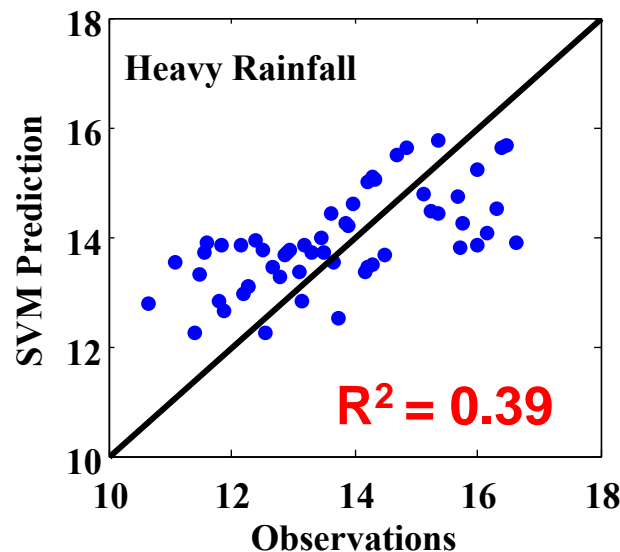
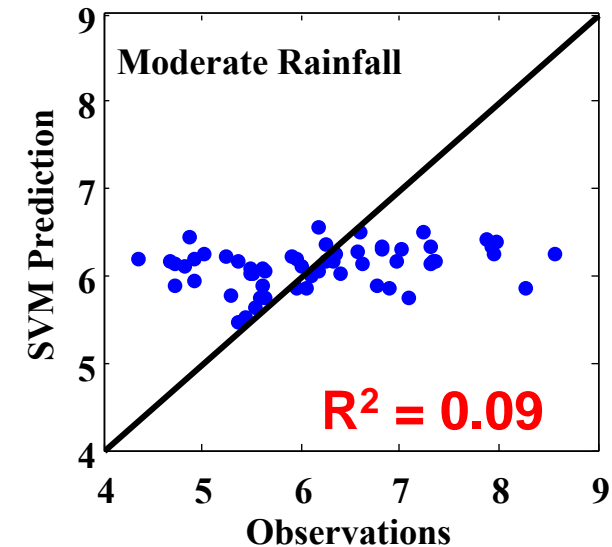
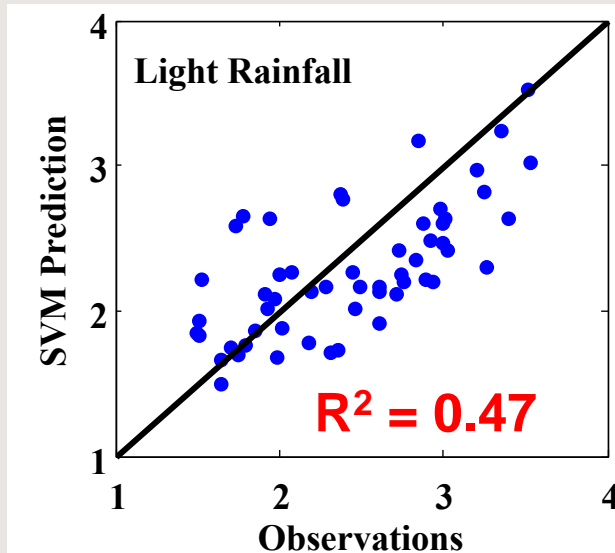
Prediction Model: SVM machine learning



SE US Summer Rainfall Predictability

SVM generate prediction skills for *light Rainfall* and *heavy rainfall* given preceding SSTA

SVM has no prediction skill for *moderate rainfall*, most likely because moderate rainfall is generated due to atmospheric internal variability

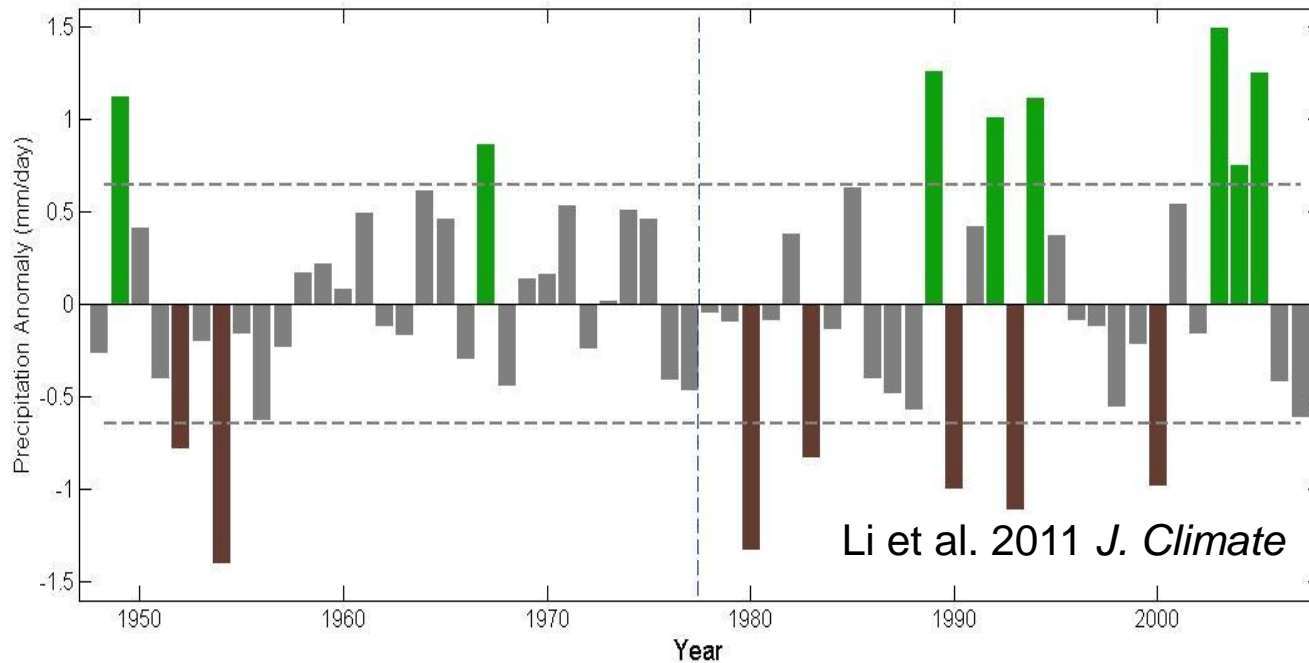


	Predictors
Light Rainfall	ENSO, Tripole SSTA (3-month lead)
Moderate Rainfall	ENSO, Tripole SSTA, Horseshoe SSTA
Heavy Rainfall	Horseshoe SSTA (2-month lead)

Concluding Remarks

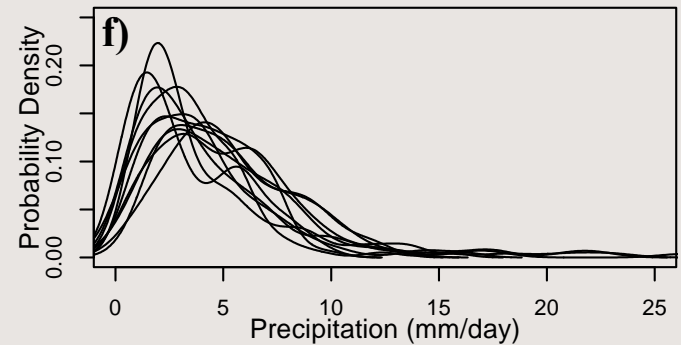
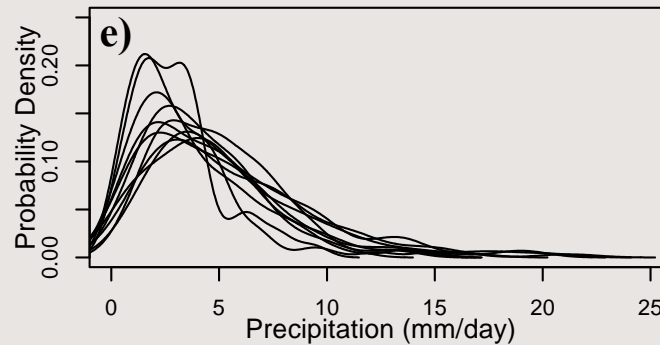
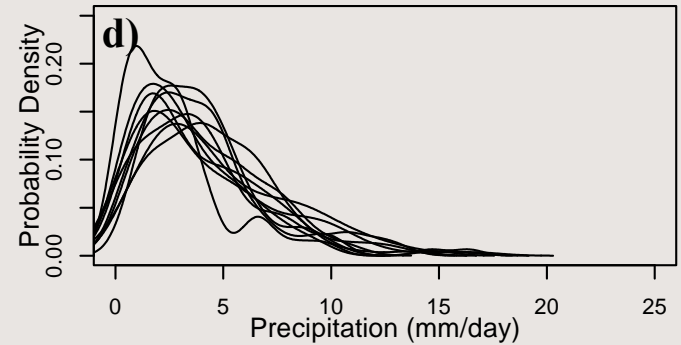
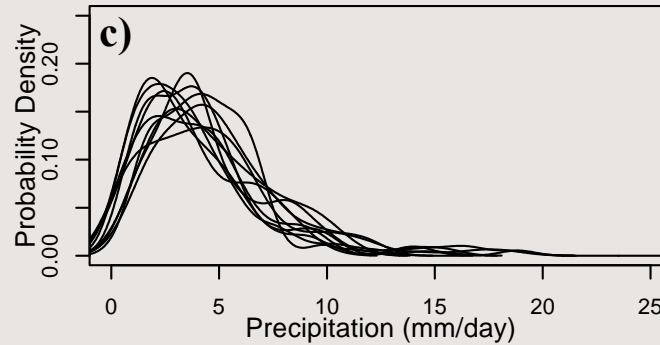
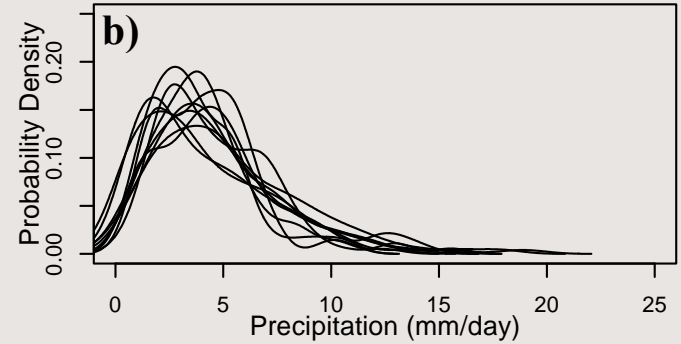
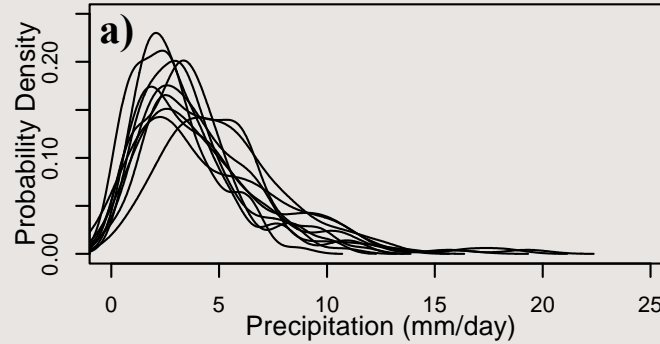
- New rainfall framework based on a 3-cluster finite *Normal* mixture model better captures the probability distribution of light, moderate, and heavy rainfall over the SE US
- 3 rainfall clusters are linked to different climate factors:
 - 1) light rainfall \leftarrow ENSO+ North Atlantic Tripole SSTA;
heavy rainfall \leftarrow Atlantic horseshoe-like SSTA
 - 2) moderate rainfall might be generated by atmospheric internal variability.
- Given the above SSTA-rainfall relationship, the support vector machine generate prediction skills in light and heavy rainfall, but no skills for moderate rainfall.

Thank You!

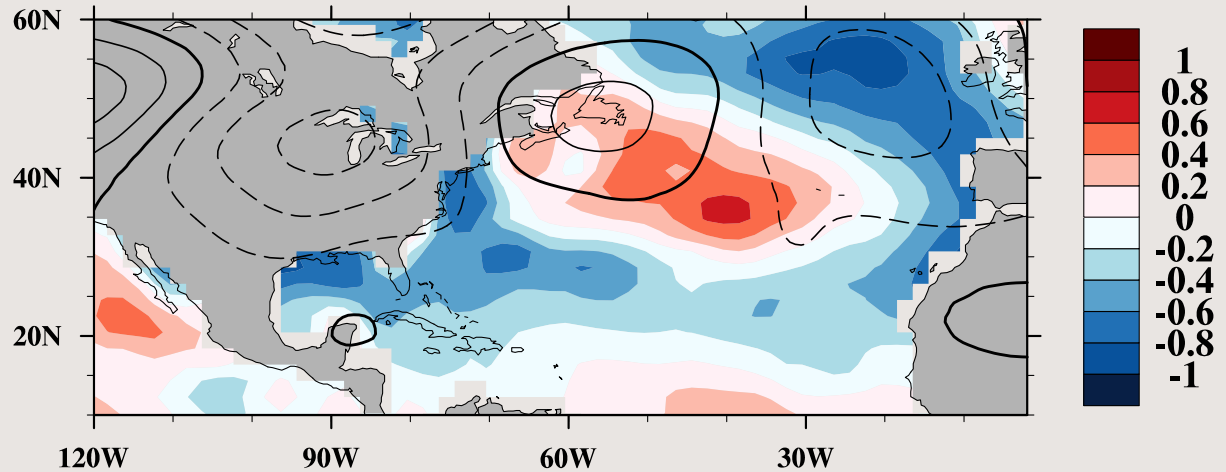
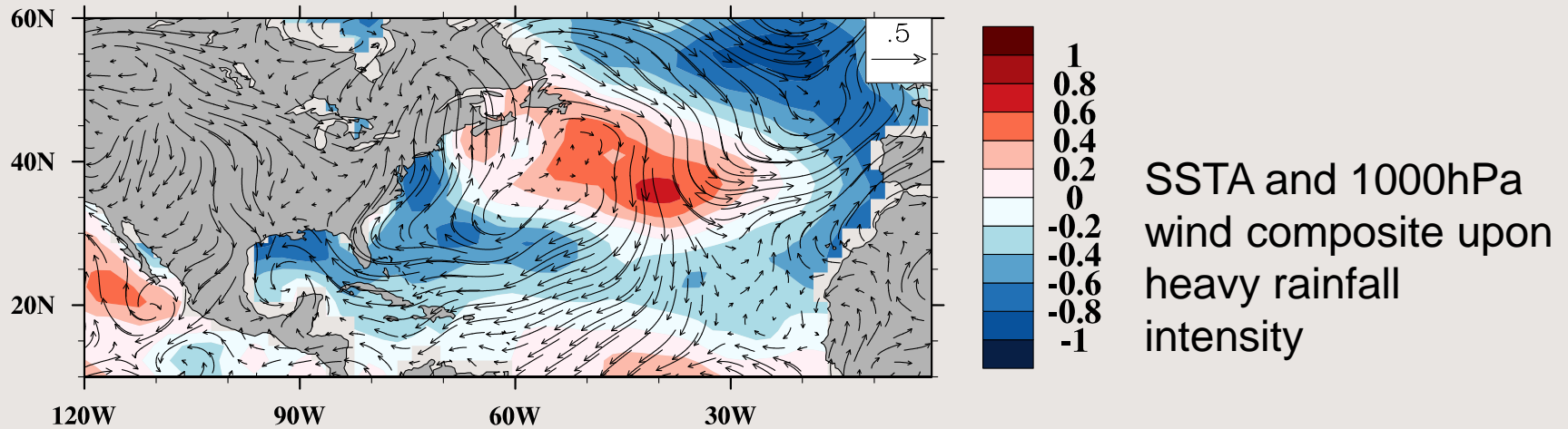


Q&A

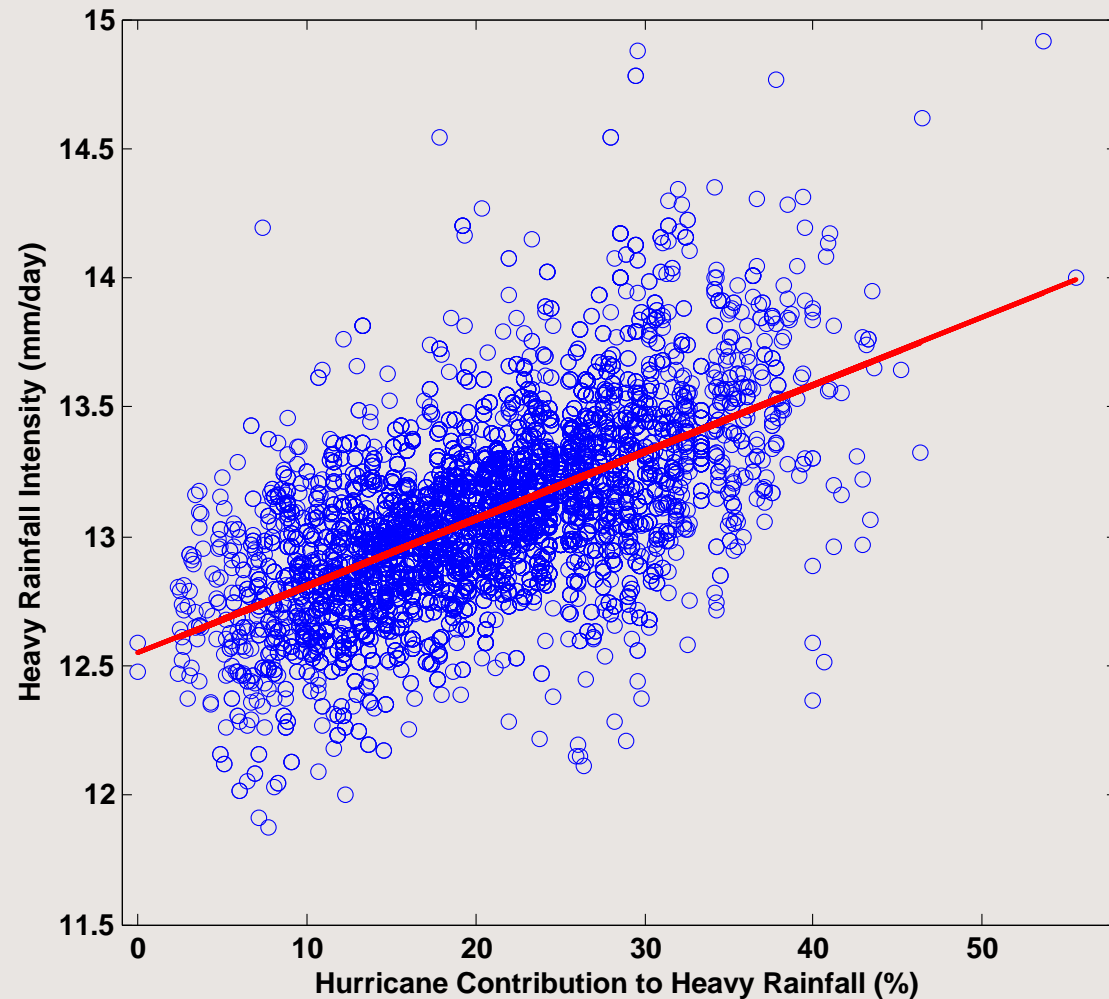
Rainfall PDFs



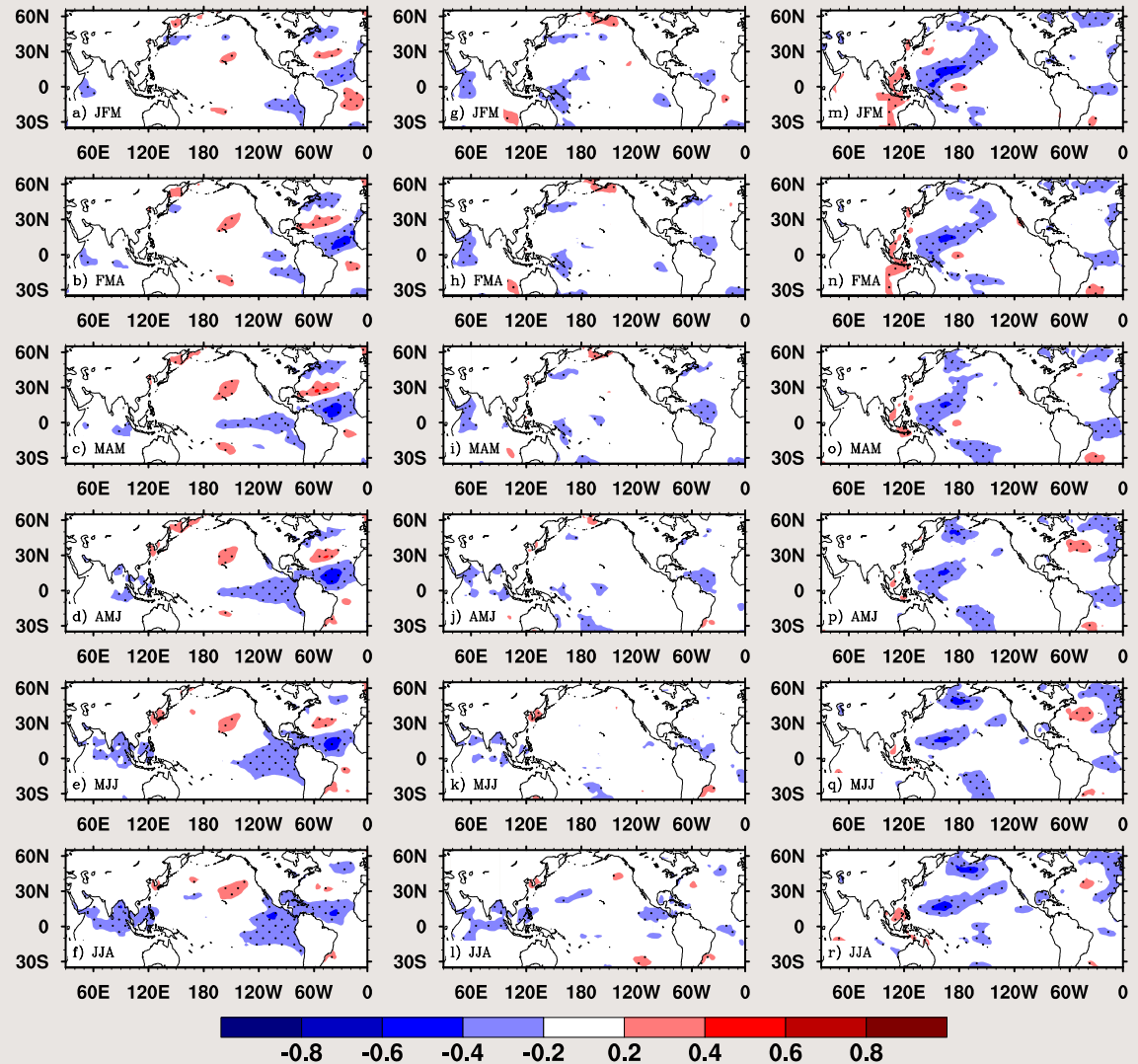
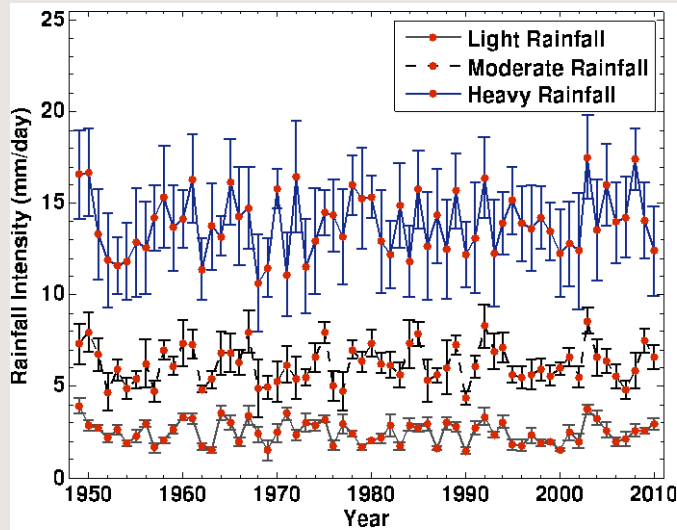
Horseshoe SSTA and Heavy Rainfall



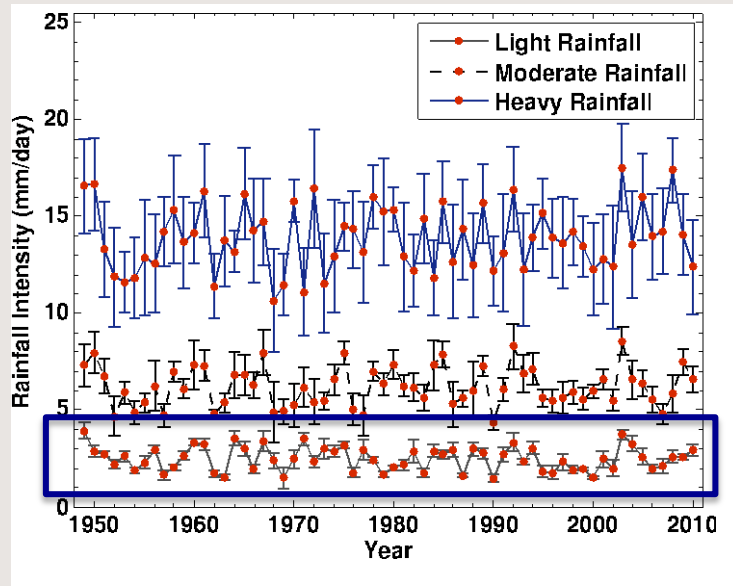
Hurricane Contribution to Heavy Rainfall



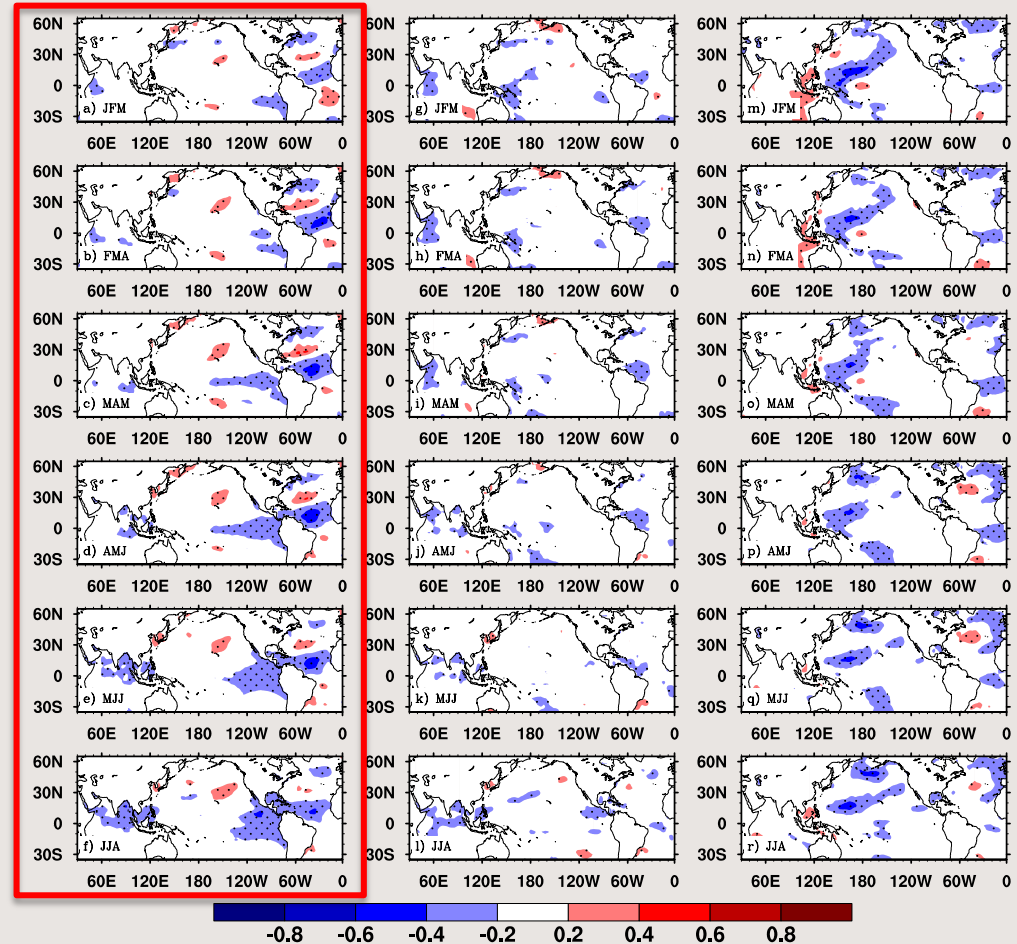
Rainfall & SSTA



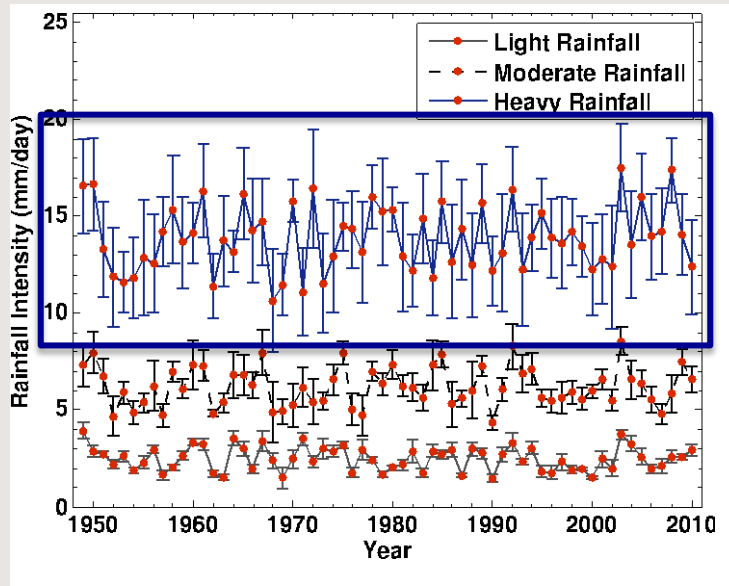
Light Rainfall & SSTA



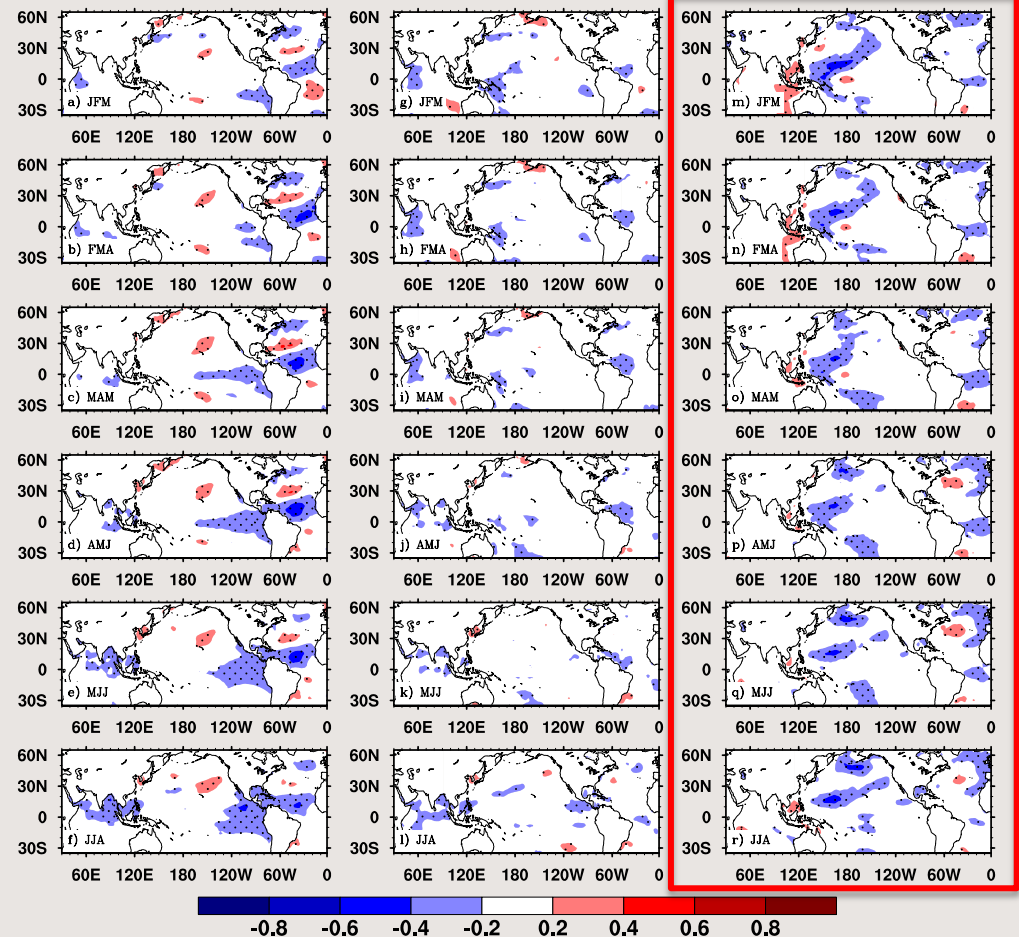
La Nina and tripole SSTA over the North Atlantic occurs 4 months ahead of the SE US light rainfall



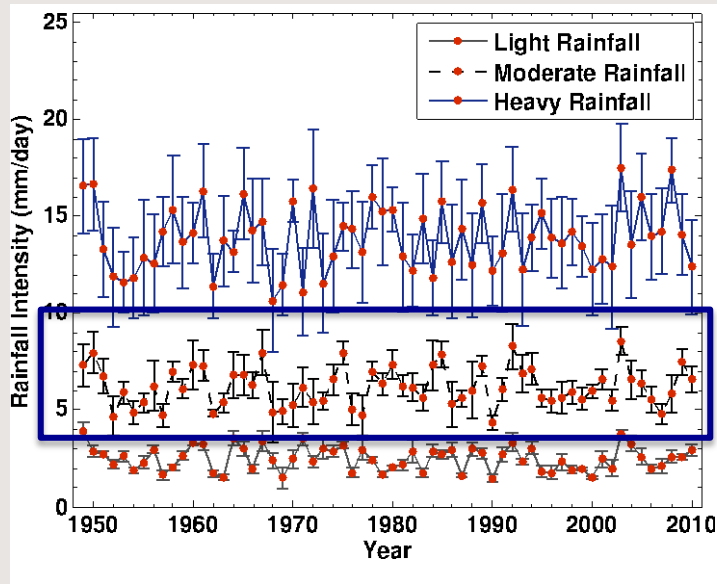
Heavy Rainfall & SSTA



SE US Heavy rainfall is associated with a “horseshoe-like” SSTA over the North Atlantic.



Moderate Rainfall & SSTA



No systematic SSTA pattern concurs with moderate rainfall

